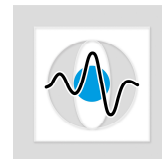


# IMIP – Exercise

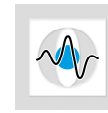
## *Epipolar Geometry & Fundamental Matrix & Normalized Eight Point Algorithm*

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Pattern Recognition Lab (CS 5)



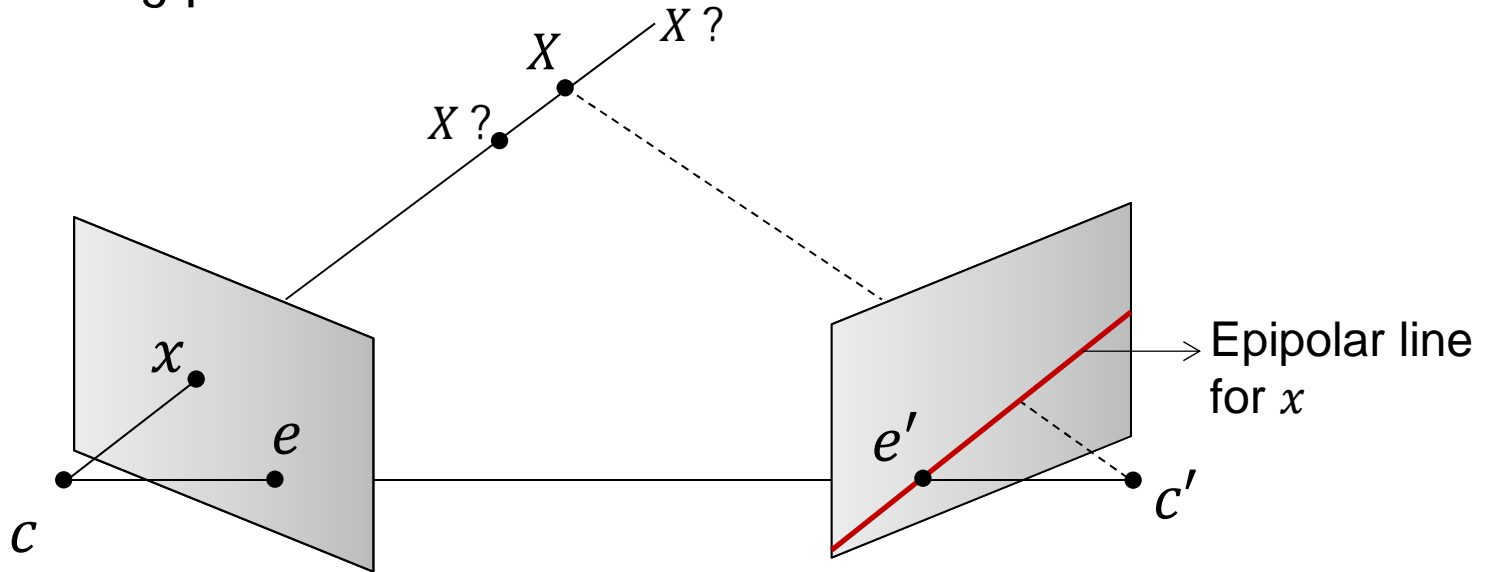
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# Epipolar Geometry

*Problem:* Given an image point in the first view, where is the corresponding point in the second view?



- A point in one image “generates” a line in the other image
- This line is known as an epipolar line, and the geometry which gives rise to it is known as epipolar geometry



## Epipolar line

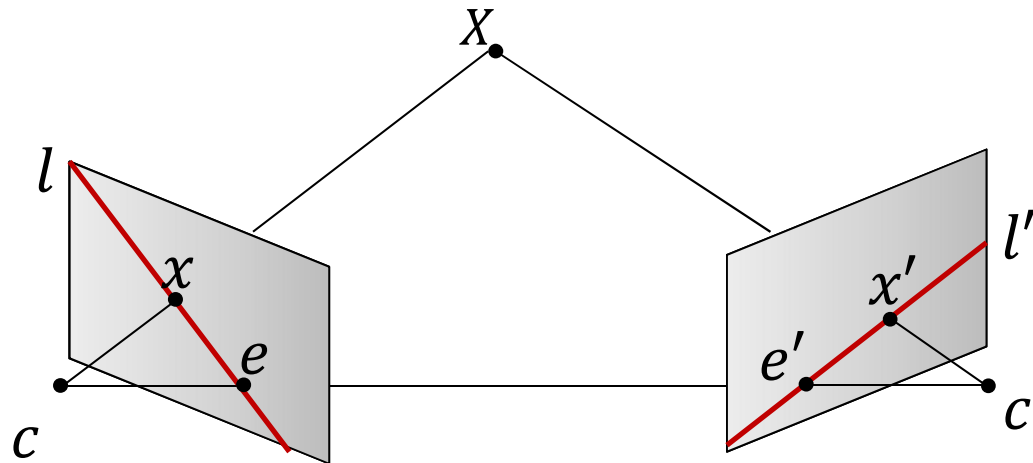


### Epipolar constraint

- Reduces correspondence problem to 1D search along an epipolar line



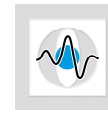
# Fundamental Matrix



- If  $x$  and  $x'$  are corresponding image points, then the fundamental matrix  $F$  is defined by

$$\mathbf{x}'^T \mathbf{F} \mathbf{x} = 0. \quad (\text{see lecture slides 11, 26})$$

- Epipolar lines:
  - $\mathbf{l}' = \mathbf{F} \mathbf{x}$  is the epipolar line corresponding to  $\mathbf{x}$
  - $\mathbf{l} = \mathbf{F}^T \mathbf{x}'$  is the epipolar line corresponding to  $\mathbf{x}'$



## Computing Fundamental Matrix

- The basic relation is

$$\mathbf{x}'^T \mathbf{F} \mathbf{x} = 0.$$

- The equation for a pair of points, e.g.,  $(x, y, 1)$  and  $(x', y', 1)$ :

$$x'x f_{11} + x'y f_{12} + x' f_{13} + y'x f_{21} + y'y f_{22} + y' f_{23} + x f_{31} + y f_{32} + f_{33} = 0$$

Gives an equation

$$(x'x, x'y, x', y'x, y'y, y', x, y, 1) \mathbf{f} = 0$$

where

$$\mathbf{f} = (f_{11}, f_{12}, f_{13}, f_{21}, f_{22}, f_{23}, f_{31}, f_{32}, f_{33})^T$$

holds the entries of the fundamental matrix.



# Computing Fundamental Matrix

- For n corresponding points

$$\mathbf{A}\mathbf{f} = \underbrace{\begin{bmatrix} x'_1x_1 & x'_1y_1 & x'_1 & y'_1x_1 & y'_1y_1 & y'_1 & x_1 & y_1 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x'_nx_n & x'_ny_n & x'_n & y'_nx_n & y'_ny_n & y'_n & x_n & y_n & 1 \end{bmatrix}}_{\text{Measurement matrix}} \begin{pmatrix} f_{11} \\ f_{12} \\ f_{13} \\ f_{21} \\ f_{22} \\ f_{23} \\ f_{31} \\ f_{32} \\ f_{33} \end{pmatrix} = 0$$



## Linear Solution

- We have a homogeneous set of equations  $Af = 0$
- $f$  can be determined only up to a scale, so there are 8 unknowns, and at least 8 point correspondences are needed  
→ hence the name “8 point algorithm”
- The least square solution is the singular vector corresponding the smallest singular values of  $A$ 
  - Take SVD:  $A = UDV^T$
  - Solution is last column of  $V$



# Constraint Enforcement

***(Enforce the rank-2 constraint)*** →

Why? See slide 15

If  $f$  is computed linearly from 8 or more correspondences, singularity condition ( $\det(\mathbf{F}) = \mathbf{0}$ ) does not hold.

## SVD Method

- (i) SVD:  $\mathbf{F} = \mathbf{U}\mathbf{D}\mathbf{V}^T$
- (ii)  $\mathbf{U}$  and  $\mathbf{V}$  are orthogonal,  $\mathbf{D} = \text{diag}(r,s,t)$
- (iii)  $r \geq s \geq t$
- (iv) Set  $\mathbf{F}' = \mathbf{U} \text{diag}(r,s,0) \mathbf{V}^T$
- (v) Resulting  $\mathbf{F}'$  is singular



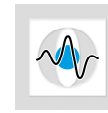


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Problem: 8 point algorithm performs badly in presence of noise

Normalization of data (balancing)

- (i) Translate so centroid is at origin
- (ii) Scale so that RMS distance of points from origin is  $\sqrt{2}$



## (i) Translate so centroid is at origin

We first compute the centroid of the set of all points:

$$t_x = \frac{1}{N} \sum_{k=1}^N p_{x,k}^i, \quad t_y = \frac{1}{N} \sum_{k=1}^N p_{y,k}^i,$$

where  $t = (t_x, t_y)$  the centroid of the  $k=1, \dots, N$  points  $p_k^i$

Then, we can center the points by

$$mc = p_k^i - t$$



## (ii) Scaling

The distance of each new computed point to the origin (0,0)

$$dc = \sqrt{\sum (mc)^2}$$

$dc$  divided by the number of tracked points, resulting in the average distance to the origin.

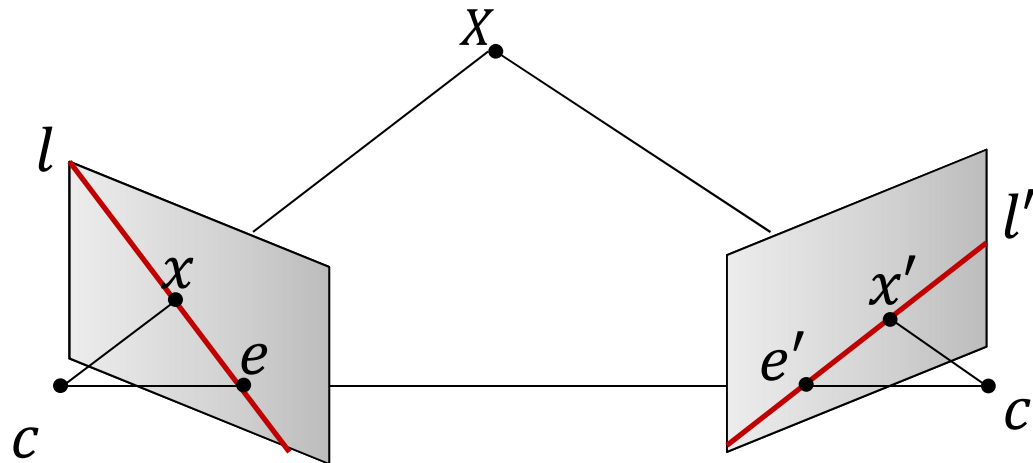
Then, we apply a scale factor  $s$ , to satisfy the criteria that the average distance of a point  $p$  from the origin is  $\sqrt{2}$ .

Finally, the normalized data can be computed by apply the transformation

$$T = \begin{pmatrix} s & 0 & -st_x \\ 0 & s & -st_y \\ 0 & 0 & 1 \end{pmatrix}$$



# Compute Epipolar line and Epipole



*Epipolar lines:*

- $\mathbf{l}' = \mathbf{F}\mathbf{x}$  is the epipolar line corresponding to  $\mathbf{x}$
- Assume  $\mathbf{l}' = (a, b, c)$ , then line equation is
$$ax + by + c = 0$$
- $\mathbf{x}' = (x', y', 1)$  lies on the line  $\mathbf{l}'$ 
$$\rightarrow 0 = ax' + by' + c = \mathbf{x}'^T \mathbf{l}'$$
$$\rightarrow 0 = \mathbf{x}'^T \mathbf{F}\mathbf{x}$$

*Epipole:  $\mathbf{F}\mathbf{e} = \mathbf{0}$  (slide 27)  $\rightarrow e$  is eigenvector with smallest eigenvalue*